Python for Data processing

Lecture 3: pandas - Part I

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What we already know

A lot about **numpy** arrays and PyTorch tensors:

- basic operations
- best practices
- optimization
- linear algebra
- very basics of how machine learning works

Structured and unstructured data

Unstructured data

- images
- signals (including time series)
- text

Each data element (pixel, datapoint, letter) is usually atomic and **is equal** to any other data element. You need to **perform analysis** to get the structure

Structured data

- tabular data
- JSON
- XML

Each data element (row, DB record, XML file) has **internal structure** or **schema**

```
[{'name': 'Anny Smith', 'age': 35, 'sex': 'female'},
{'name': 'John Black', 'age': 62, 'sex': 'male'}, ...]
```

Dataframe

Tabular representation of structured data

- well known in R world for years
- **indexed** rows and columns
- SQL-like operations^(joins, filtering), aggregations, alignment and more

pandas

One of the most respected Python packages for data science

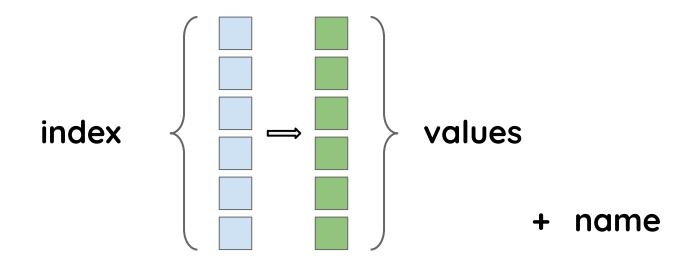
- started in 2008
- **very fast** (a lot of Cython inside)
- supports tons of operations and formats
- extremely flexible and powerful
- It's **crazy** sometimes, but you'll **love** it

pandas series and df's

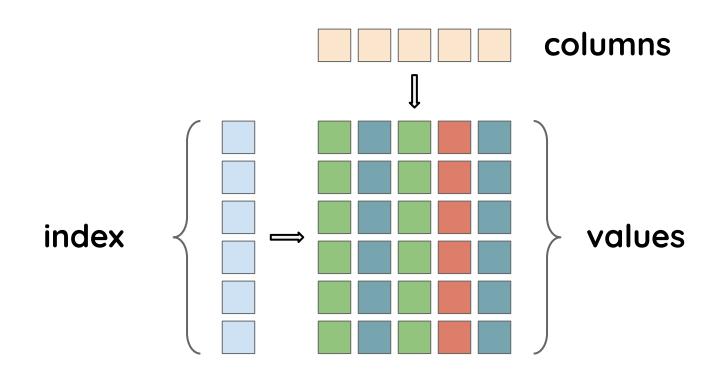
pandas has two main data structures:

- pd.Series for indexed 1D data
- pd.DataFrame for indexed tabular data

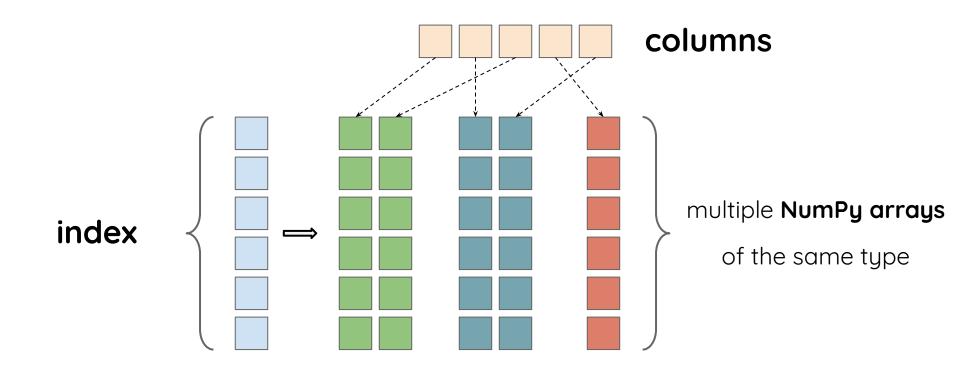
pandas series



pandas dataframe



pandas dataframe



Indexing series and dataframes

Indexing series and df's

- [] indexing
- .loc label based indexing
- .iloc label based indexing

Boolean indexing is possible and is heavily used

Indexing df: SettingWithCopyWarning

pandas is not like numpy:

- It's unknown whether you get view or copy

Why?

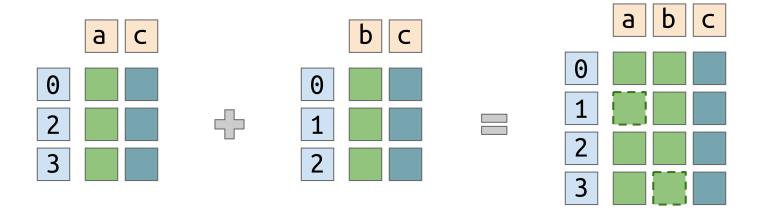
- It's hard to give guarantees
(but there are rules inside) (but you should not even try to understand them)

Operations on dataframes

Arithmetic: not your usual numpy

pandas aligns dataframes for you before performing operations by creating a **union** of row and column indexes

Arithmetic



Applying functions to df's

pandas allows you to apply custom functions across rows or columns, and elementwise.

And combines results for you appropriately.

It's usually **fast**.

Reading csv files

The best tool to read CSV and other text files in Python:

pd.read_csv(...)

Dataframe summaries

```
It's easy to get general information about dataframe:

df.info()

df.describe()

df.head(), df.tail()
```

Counts and statistics

To get counts or statistics about column or row:

df[col].unique()

df[col].value_counts()

df.sum(axis=...), df.mean(axis=...), df.std(axis=...)

Powerful in combination with smart indexing.

Groupby operations (basic)

Dataframes can be **grouped** by column or columns, or row index level. Great for:

- calculating fine grained statistics
- plotting
- applying operations group-wise

.groupby is an extremely powerful tool

Replacing and renaming

df.replace allows for flexible replacement of values in dataframes:

- by value, per column

df.rename allows you to easily rename any label, be it column name or index label

Missing data

pandas is great at handling missing data:

- infers it for you
- backward fill, forward fill and more

Categorical data

pandas easily calculates one-hot encoded values for any column, adding properly named columns

Special datatypes

pandas has very good support for:

- **strings** great for text columns (split, replace and other usual string operations, vectorized)
- datetime flexible indexing, handling timezones and extravagant parsing (great for anything time series related)

What we've learned

Basics of **pandas**:

- creating, indexing
- operations on dataframes
- basics of grouping

Words we know

- Jupyter
- numpy
- PyTorch
- pandas

Assignment

- explore pandas
- play with Titanic dataset

questions?